

**Review.** If  $v_1, \dots, v_n$  are orthogonal, the orthogonal projection of  $w$  onto  $\text{span}\{v_1, \dots, v_n\}$  is

$$\hat{w} = \frac{w \cdot v_1}{v_1 \cdot v_1} v_1 + \dots + \frac{w \cdot v_n}{v_n \cdot v_n} v_n.$$

**Example 69.**

(a) Project  $\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$  onto  $W = \text{span}\left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} \right\}$ .

(b) Express  $\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$  in terms of the basis  $\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ -5 \end{bmatrix}$ .

**Solution.**

(a) We note that the vectors  $\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix}$  are orthogonal to each other.

Therefore, the projection can be computed as  $\frac{\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}}{\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + \frac{\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix}}{\begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix}} \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} = \frac{8}{6} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + \frac{4}{5} \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix}$ .

**Comment.** If we didn't have an orthogonal basis for  $W = \text{col}\left(\begin{bmatrix} 1 & 2 \\ 2 & -1 \\ 1 & 0 \end{bmatrix}\right)$ , then we would have to solve the least squares problem  $\begin{bmatrix} 1 & 2 \\ 2 & -1 \\ 1 & 0 \end{bmatrix} x = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$  instead to get the same final result (with more work).

(b) Note that this basis is orthogonal! Therefore, we can compute  $\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix} = \frac{8}{6} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + \frac{4}{5} \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix} + \frac{5}{30} \begin{bmatrix} 1 \\ 2 \\ -5 \end{bmatrix}$ .

(We proceed exactly as in the previous part to compute each coefficient as a quotient of dot products.)

## Gram-Schmidt

### (Gram-Schmidt orthogonalization)

Given a basis  $w_1, w_2, \dots$  for  $W$ , we produce an orthogonal basis  $q_1, q_2, \dots$  for  $W$  as follows:

- $q_1 = w_1$
- $q_2 = w_2 - \left( \begin{matrix} \text{projection of} \\ w_2 \text{ onto } q_1 \end{matrix} \right)$
- $q_3 = w_3 - \left( \begin{matrix} \text{projection of} \\ w_3 \text{ onto } q_1 \end{matrix} \right) - \left( \begin{matrix} \text{projection of} \\ w_3 \text{ onto } q_2 \end{matrix} \right)$
- $q_4 = \dots$

**Note.** Since  $q_1, q_2$  are orthogonal,  $\left( \begin{matrix} \text{projection of} \\ w_3 \text{ onto } \text{span}\{q_1, q_2\} \end{matrix} \right) = \left( \begin{matrix} \text{projection of} \\ w_3 \text{ onto } q_1 \end{matrix} \right) + \left( \begin{matrix} \text{projection of} \\ w_3 \text{ onto } q_2 \end{matrix} \right)$ .

**Important comment.** When working numerically on a computer it actually saves time to compute an orthonormal basis  $q_1, q_2, \dots$  by the same approach but always normalizing each  $q_i$  along the way. The reason this saves time is that now the projections onto  $q_i$  only require a single dot product (instead of two). This is called **Gram-Schmidt orthonormalization**. When working by hand, it is usually simpler to wait until the end to normalize (so as to avoid working with square roots).

**Note.** When normalizing, the orthonormal basis  $q_1, q_2, \dots$  is the unique one (up to  $\pm$  signs) with the property that  $\text{span}\{q_1, q_2, \dots, q_k\} = \text{span}\{w_1, w_2, \dots, w_k\}$  for all  $k = 1, 2, \dots$

**Example 70.** Using Gram–Schmidt, find an orthogonal basis for  $W = \text{span}\left\{\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}\right\}$ .

**Solution.** We already have the basis  $\mathbf{w}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ ,  $\mathbf{w}_2 = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$  for  $W$ . However, that basis is not orthogonal.

We can construct an orthogonal basis  $\mathbf{q}_1, \mathbf{q}_2$  for  $W$  as follows:

- $\mathbf{q}_1 = \mathbf{w}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$
- $\mathbf{q}_2 = \mathbf{w}_2 - \left(\begin{array}{c} \text{projection of} \\ \mathbf{w}_2 \text{ onto } \mathbf{q}_1 \end{array}\right) = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} - \frac{1}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 2 \\ -4 \\ 2 \end{bmatrix}$

**Note.**  $\mathbf{q}_2$  is the error of the projection of  $\mathbf{w}_2$  onto  $\mathbf{q}_1$ . This guarantees that it is orthogonal to  $\mathbf{q}_1$ . On the other hand, since  $\mathbf{q}_2$  is a combination of  $\mathbf{w}_2$  and  $\mathbf{q}_1$ , we know that  $\mathbf{q}_2$  actually is in  $W$ .

We have thus found the orthogonal basis  $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \frac{2}{3} \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$  for  $W$  (if we like, we can, of course, drop that  $\frac{2}{3}$ ).

**Important comment.** By normalizing, we get an orthonormal basis for  $W$ :  $\frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$ .

**Practical comment.** When implementing Gram–Schmidt on a computer, it is beneficial (slightly less work) to normalize each  $\mathbf{q}_i$  during the Gram–Schmidt process. This typically introduces square roots, which is why normalizing at the end is usually preferable when working by hand.

**Comment.** There are, of course, many orthogonal bases  $\mathbf{q}_1, \mathbf{q}_2$  for  $W$ . Up to the length of the vectors, ours is the unique one with the property that  $\text{span}\{\mathbf{q}_1\} = \text{span}\{\mathbf{w}_1\}$  and  $\text{span}\{\mathbf{q}_1, \mathbf{q}_2\} = \text{span}\{\mathbf{w}_1, \mathbf{w}_2\}$ .

A matrix  $Q$  has orthonormal columns  $\iff Q^T Q = I$

**Why?** Let  $\mathbf{q}_1, \mathbf{q}_2, \dots$  be the columns of  $Q$ . By the way matrix multiplication works, the entries of  $Q^T Q$  are dot products of these columns:

$$\begin{bmatrix} - & \mathbf{q}_1^T & - \\ - & \mathbf{q}_2^T & - \\ & \vdots & \end{bmatrix} \begin{bmatrix} | & | & \dots \\ \mathbf{q}_1 & \mathbf{q}_2 & \dots \\ | & | & \dots \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & \ddots \end{bmatrix}$$

Hence,  $Q^T Q = I$  if and only if  $\mathbf{q}_i^T \mathbf{q}_j = 0$  (that is, the columns are orthogonal), for  $i \neq j$ , and  $\mathbf{q}_i^T \mathbf{q}_i = 1$  (that is, the columns are normalized).

**Example 71.**  $Q = \begin{bmatrix} 1/\sqrt{3} & 1/\sqrt{6} \\ 1/\sqrt{3} & -2/\sqrt{6} \\ 1/\sqrt{3} & 1/\sqrt{6} \end{bmatrix}$  obtained from Example 70 satisfies  $Q^T Q = I$ .